Multi-modal albedo distributions in the ablation area of the southwestern Greenland Ice Sheet

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1   Abstract

2   Surface albedo is a key variable controlling solar radiation absorbed at the Greenland Ice

3   Sheet (GrIS) surface, and thus, meltwater production. Recent decline in surface albedo over

4   the GrIS has been linked to enhanced snow grain metamorphic rates, earlier snowmelt, and

5   amplified ice-albedo feedback from atmospheric warming. However, the importance of

6   distinct surface types on ablation area albedo and meltwater production is still relatively

7   unknown. In this study, we analyze albedo and ablation rates using in situ and remotely-

8   sensed data. Observations include: 1) a new high-quality in situ spectral albedo dataset

9   collected with an Analytical Spectral Devices (ASD) spectroradiometer measuring at 325–

10   1075 nm, along a 1.25 km transect during three days in June 2013; 2) broadband albedo at

11   two automatic weather stations; and 3) daily MODerate Resolution Imaging

12   Spectroradiometer (MODIS) albedo (MOD10A1) between 31 May – 30 August 2012 and

13   2013. We find that seasonal ablation area albedos in 2013 have a bimodal distribution, with

14   snow and ice facies characterizing the two peaks. Our results show that a shift from a

15   distribution dominated by high to low albedos corresponds to an observed melt rate increase

16   of 51.5% (between 10 – 14 July and 20 – 24 July, 2013). In contrast, melt rate variability

17   caused by albedo changes before and after this shift was much lower, and varied between

18   ~10-30% in the melting season. Ablation area albedos in 2012 exhibited a more complex

19   multimodal distribution, reflecting a transition from light to dark-dominated surface, as well

20   as sensitivity to the so called 'dark band' region in southwest Greenland. In addition to a

21   darkening surface from ice crystal growth, our findings demonstrate that seasonal changes in

22   GrIS ablation area albedo are controlled by changes in the fractional coverage of snow, bare

23   ice, and impurity-rich surface types. Thus, seasonal variability in ablation area albedo appears

24   to be regulated primarily as a function of bare ice expansion at the expense of snow, surface

25   meltwater ponding, and melting of outcropped ice layers enriched with mineral materials,

26   enabling dust and impurities to accumulate. As climate change continues in the Arctic region,

27   understanding the seasonal evolution of ice sheet surface types in Greenland's ablation area is

28   critical to improve projections of mass loss contributions to sea level rise.

29

30   1   Introduction

31   Surface albedo, defined as the bihemispherical reflectance integrated across the visible and

32   near-infrared wavelengths (Schaepman-Strub et al., 2006), is a key variable controlling

1   Greenland Ice Sheet (GrIS) surface melting. During the melt season, surface albedo
2   modulates absorbed solar radiation at the ice surface, and consequently, the surface energy
3   and mass balance of the ice sheet (Cuffey and Paterson, 2010). Over the last decade, an
4   observed decline in albedo has been linked to less summer snow cover, expansion of bare ice
5   area, and enhanced snow grain metamorphic rates from atmospheric warming, amplified by
6   the melt-albedo feedback (Box et al., 2012; Stroeve et al., 2013; Tedesco et al., 2011). This
7   positive feedback involves increased melting, and exposure of bare ice, impurities and
8   meltwater ponding, reducing surface albedo, thereby increasing solar radiation absorption,
9   and thus, accelerating melt further (Box et al., 2012; Tedesco et al., 2011).

10   The GrIS surface has a wide range of surface types with different albedos, including
11   snow, ice, dust and sediment-rich impurities, cryoconite holes, melt ponds, and streams. Yet,
12   the importance of these surface types on ablation area albedo, and thus, meltwater production
13   over the melt season is still relatively unresolved (Rennermalm et al., 2013). Current state-of-
14   the-art surface mass balance (SMB) models, such as Modèle Atmosphérique Régionale
15   (MAR) v3.2 and Regional Atmospheric Climate MOdel (RACMO2), consider some
16   variability in surface types by including the presence of meltwater ponding, snow, and bare
17   ice surfaces to characterize seasonal variations in ablation area albedo (Alexander et al., 2014;
18   Van Angelen et al., 2012). Furthermore, RACMO2 considers the presence of black carbon
19   concentrations on snow and is capable of utilizing realistic MODIS background albedo data
20   (Van Angelen et al., 2012), thereby representing the impact of surface types spatially
21   aggregated to the MODIS resolution. However, few studies have utilized these modeling tools
22   to understand how the distributions of surface types are changing ablation area albedo (e.g.,
23   Alexander et al. 2014). This is increasingly important due to enhanced surface melt associated
24   with anomalously warm atmospheric circulation patterns in 2007-2012 that may become more
25   frequent in the future (Hall et al., 2013; Nghiem et al., 2012; Tedesco et al., 2013).
26   Additionally, some studies suggest that a new control of ice sheet albedo is the deposition and
27   accumulation of light-absorbing impurities advected from snow-free areas and forest fires
28   outside of Greenland (Dumont et al., 2014; Keegan et al., 2014).

29   The large-scale decline in albedo has been greatest in southwest Greenland (-0.04 to -
30   0.16 per decade trend in June and August, respectively; Stroeve et al., 2013). This is related to
31   stronger warming trends (2-4 °C in some regions; Hanna et al., 2014), early melt onset, a lack
32   of wintertime accumulation (van den Broeke et al., 2008), expansion of bare ice area (Tedesco

et al., 2011), high concentration of impurities (cryoconite, dust, and soot), melting of outcropped ice layers enriched with mineral content (Wientjes and Oerlemans, 2010; Wientjes et al., 2011), and enhanced meltwater production and runoff (e.g., Mernild et al., 2012). Seasonal changes in the distribution of different surface types in southwest Greenland's ablation area have considerable influence on the spatiotemporal variability of surface albedo (Chandler et al., 2014; Knap and Oerlemanns, 1996; Konzelmann and Braithwaite, 1995). During the melt season, surface albedo decreases as cryoconite hole coverage increases (Chandler et al., 2014), melt ponds and supraglacial rivers form efficient drainage networks (Lampkin and VanderBerg, 2013; Kang and Smith, 2013; Smith et al., 2015), crevasses and other types of roughness begin to form, and impurities accumulate from exposure of the underlying ice surface (Wientjes and Oerlemans, 2010). Albedo in western Greenland's ablation area averages around ~0.41 for the duration of the melt season (Wientjes et al., 2011), but can vary between > 0.80 for fresh snow, to 0.30-0.60 for bare ice (Cuffey and Patterson, 2010), and ~0.10 for cryoconite surfaces (Bøggild et al., 2010; Chandler et al., 2014; Knap and Oerlemans, 1996). Furthermore, negative albedo trends since 2000 (Box et al., 2012) are linked to an expansion of areas of ablation relative to accumulation facies.

Changes in surface albedo are typically characterized from the MODerate Resolution Imaging Spectroradiometer (MODIS) and the Advanced Very High Resolution Radiometer (AVHRR) satellite sensors (e.g., Chandler et al., 2014; Stroeve et al., 2013; Wang et al., 2012; Wright et al., 2014) or modeled with regional climate models (RCMs) such as Regional Atmospheric Climate Model (RACMO2; Van Meijgaard et al., 2008) and Modèle Atmosphérique Régional (MAR; Fettweis, 2007). Remotely-sensed and modeled albedo has been validated with ground measurements from dispersed Greenland Climate Network Automatic Weather Stations (GC-Net AWS; Knap and Oerlemans, 1996; Steffen and Box, 2001). These comparisons reveal that satellite products provide reasonable albedo estimates (Box et al., 2012; Stroeve et al., 2005, 2006, 2013), although discrepancies between different MODIS albedo products have been identified (Alexander et al., 2014). Despite this, RCM surface albedos remain represented in relatively simplistic terms, particularly in regions that frequently experience prolonged bare ice exposure like southwest Greenland (Fettweis et al, 2011; Fitzgerald et al., 2012; Rae et al., 2012; Van Angelen et al., 2012). This is attributed to a lack of surface roughness in the RCMs (Ettema et al., 2010), and relatively simplistic bare ice and impurity albedo schemes (Alexander et al., 2014), resulting in large inter-model differences in runoff (42% variance; Vernon et al., 2013), despite the existence of spatially

distributed ice albedo schemes and inclusion of black carbon contaminants on snow surfaces (Van Angelen et al., 2012). Recent surface albedo observations and snow model simulations of impurity-rich surfaces have been linked to enhanced ice sheet melt (Chandler et al., 2014; Dumont et al., 2014; Keegan et al., 2014), suggesting that incorporating seasonal changes in the albedo distribution of distinct surface types might improve accuracy of modeled meltwater runoff and GrIS sea level rise contributions. These findings point to the importance of a detailed assessment of high spectral, spatial, and temporal resolution albedo data to quantify how different surface types control ablation area albedo, and therefore, melt.

In this study, we report the results of an assessment of ablation area albedo along the southwestern GrIS for the 2012 and 2013 melt seasons. We use 1) a new high-quality in situ spectral albedo dataset collected with an Analytical Spectral Devices (ASD) spectroradiometer measuring over a wavelength range of 325–1075 nm, along a 1.25 km transect during three days in June 2013; 2) in situ broadband albedo at two automatic weather stations; and 3) daily MODerate Resolution Imaging Spectroradiometer (MODIS) albedo (MOD10A1) product (Hall et al., 2012) between 31 May – 30 August 2012 and 2013 to investigate how ice sheet surface types influence surface albedo and ablation rates; and 4) summer seasonal changes in surface type coverage reported in literature. First, we describe the collection of high-quality in situ spectral albedo, automatic weather station broadband albedo, and ablation stake measurements collected during early 2013 melt season along a fixed transect in the GrIS ablation area. Second, from the MODIS daily albedo data, we estimate seasonal changes in the albedo distributions. These distributions were compared with seasonal changes in computed albedo distributions derived by using in situ and literature values of albedo for distinct surface type and fractional area of surface types from a nearby site (1030 m a.s.l.; reported by Chandler et al., 2014). Thirdly, the impact of changing albedo and surface type coverage on surface melt was quantified and compared with transect ablation stake measurements. Finally, we compare these results with 2012 MOD10A1 data to better understand the overall frequency distribution, spatiotemporal variability, and ablation rates associated with dominant surface types in southwest Greenland's ablation area. This study is the first high spatial, temporal, and spectral resolution albedo dataset collected in the southwestern GrIS ablation area.

2   Study site description

The study site is located on the southwestern GrIS approximately 30 km northeast of Kangerlussuaq, Greenland (Fig. 1). Albedo measurements were collected along a 1.25 km transect situated between ~510 to 590 m a.s.l., well within the ablation area for this region (mean equilibrium line altitude of 1553 m a.s.l.; van de Wal et al., 2012). Two meteorological stations, referred to as Base and Top Met Stations, were installed near the transect end points by Site E and A, respectively (Fig. 1) to derive independent measurements of in situ broadband albedo (300-1100 nm), hereafter α_{base} and α_{top}. In addition, ablation stakes were installed at five sites along the albedo transect and by the Base Met Station to measure ice surface ablation rates. Ice sheet surface types examined included white ice, shallow supraglacial streams, and dirty ice, where dirty ice was qualitatively distinguished from white ice based on visible surface sediments. Visual assessment in the study area revealed that snow had melted before mid-June and no snowfall events occurred between 8-26 June 2013. A few small melt ponds (< 1 km^2) were observed in the study area, but likely not in sufficient quantity to explain discrepancies between in situ and MODIS albedo-derived estimates.

3   Methods

3.1   Field spectroscopy measurements

High spatial (~10 m posting), temporal (1-2 days), and spectral (1 nm) resolution spectral albedo measurements, hereafter α_{ASD}, were measured at 325-1075 nm using an ASD Fieldspec HandHeld 2 Spectroradiometer (PANalytical, formerly ASD Inc.). The ASD was mounted on a tripod at 0.4 m distance, fitted with a Remote Cosine Receptor (RCR) foreoptic (with no other foreoptic attached; i.e., bare fiber), and had a 25° field-of-view, corresponding to a spot size of ~0.18 m diameter on the surface.

Spectral albedos were measured along the transect starting at Site E and ending at Site A on 16, 17, 19, 21, 24, and 25 June, 2013 between 1000 and 1800 local time (1200 – 2000 GMT). After rigorous quality control (see Appendix A), only transect observations made on the 16, 19, and 25 June were used in analyses. Broadband α_{ASD} were calculated as a weighted average based on their spectral response curve and the amount of incoming solar radiation

1 over the entire spectral range at each site along the transect. These measurements were
2 compared with MOD10A1 and meteorological station data, as described in section 3.3.

3

3.2 Continuous broadband albedo measurements at meteorological stations

5 Two AWSs, the Top and Base Met Stations, were installed at each end of the transect to
6 independently measure broadband (300-1100 nm) albedo from 8-26 June 2013, hereafter α_{base}
7 and α_{top}. Daily average broadband albedo was computed using shortwave flux measured at
8 SZAs < 70° (Stroeve et al., 2005) to minimize the cosine response error inherent to the
9 pyranometers (uncertainty increases by ± 5% for SZAs > 70°; Onset Computer Corp., 2010).
10 Expected accuracy of α_{base} and α_{top} is ± 10% based on the intrinsic accuracy and cosine
11 response error of the pyranometers. Although surface roughness effects on measured surface
12 albedo (e.g., Lhermitte et al., 2014) were not quantified here, analyses of α_{top} measurements
13 suggest that they were compromised by these effects, and thus, α_{base} alone is used for most
14 analyses. See Appendix A for details on surface installation conditions and tilt uncertainty
15 estimates.

16

3.3 MODIS albedo data

18 Daily MODIS broadband albedo (300-3000 nm) was acquired from the MOD10A1 product
19 (Version 005) from NASA's Terra satellite (Hall et al., 2006; Klein and Stroeve, 2002). High-
20 quality flagged MOD10A1 albedo data (periods of high SZA and cloudiness were excluded;
21 Schaaf et al., 2011) from 31 May to 30 August 2012 and 2013 (when SZAs are minimized;
22 e.g., Box et al., 2012) were used in two analyses. First, MOD10A1 albedo for pixels
23 overlapping with our transect site (Fig. 1), hereafter $\alpha_{MOD\ Pixel\ 1}$ and $\alpha_{MOD\ Pixel\ 2}$, were
24 compared with observations as described below. Second, distributions of MOD10A1 albedo
25 were examined at three spatial extents as described in section 3.5.

26 Broadband $\alpha_{MOD\ Pixel\ 1}$ and $\alpha_{MOD\ Pixel\ 2}$ were compared with α_{ASD} and α_{base}. Direct
27 comparison of α_{ASD}, α_{base}, and α_{MOD} absolute values are not possible due to different
28 wavelength ranges, and α_{MOD} is expected to have lower values than the other two datasets.
29 However, relative comparisons of spatial and temporal patterns are reasonable, because the

1   α_{MOD} is dominated by the ASD visible and near-infrared (i.e., 325-1075 nm) wavelengths. In

2   a standard Top-of-Atmosphere solar irradiance reference spectrum, the 325-1075 nm range

3   comprises 80.52% of the total irradiance in the 300-3000 nm range. The dominance of

4   reflectance in the ASD visible and near-infrared wavelengths in determining broadband

5   albedo means that α_{MOD} can be used qualitatively to provide spatiotemporal context. High-

6   quality broadband (325-1075 nm) α_{ASD} data within pixels 1 and 2, hereafter $\alpha_{ASD\ Pixel\ 1}$ and

7   $\alpha_{ASD\ Pixel\ 2}$, were averaged together to indirectly validate $\alpha_{MOD\ Pixel\ 1}$ and $\alpha_{MOD\ Pixel\ 2}$ data, and to

8   facilitate comparison between in situ and remotely-sensed observations. While absolute

9   values will differ between the datasets, and issues of MODIS pixel separability may exist due

10 to off-nadir footprint effects (Dozier et al., 2008), the difference shouldn't change spatial and

11 temporal patterns.

12

3.4   Ablation and albedo at dominant surface types

14 Surface melting between 8 – 26 June was estimated using ablation stakes installed at the Base

15 Met Station, hereafter M_{base}, and at five sites across the albedo transect, hereafter $M_{stakeXY}$,

16 where X denotes Sites A-E, and Y denotes surface type - white ice (W), dirty ice (D), or

17 shallow 5-10 cm deep streams (S) (Fig. 1). Bamboo poles were used as stakes (Hubbard and

18 Glasser, 2005), and ablation rates were recorded every 1-3 days by measuring the distance

19 between the bamboo pole top and ice sheet surface at cm-scale resolution.

20   α_{ASD} spectra were made within 30 m of ablation stakes to identify representative surface

21 type albedos. With the exception of Site D, all sites were relatively homogenous. At Site D,

22 the two surface types could be classified into distinct groupings: clean and dirty ice. Albedos

23 of clean ice at Sites A-C, and E, hereafter α_{ASD_AW}, α_{ASD_BW}, α_{ASD_CW}, and α_{ASD_EW}, were

24 estimated by averaging broadband α_{ASD} observations made within 30 m of stakes for each

25 transect date. At Site D, albedos of clean and dirty ice, hereafter α_{ASD_DW} and α_{ASD_DD}, were

26 estimated from the histograms of α_{ASD} observations made within 30 m of stakes for each

27 transect date. At the M_{base} stake, no spectral α_{ASD} observations were made. Instead, α_{ASD_DD} is

28 assumed to be representative of albedo at the Base Met Station, hereafter α_{MET_base}. Stream

29 albedo, hereafter α_{stream}, was determined from occasional α_{ASD} measurements at various

30 shallow surface streams between 13-25 June. Cryoconite hole albedo, hereafter α_{cryo}, was

1 parameterized using published values (from Bøggild et al., 2010) of broadband albedo
2 averaged together for damp cryoconite material and cryoconite basin surface types under
3 clear-sky and overcast conditions.

4

3.5 Melt season albedo distributions

6 Two types of melt season albedo distributions were constructed: 1) computed distributions
7 based on broadband α_{ASD} for distinct surfaces and fractional surface coverage area from
8 Chandler et al. (2014); and 2) observed MODIS-derived distributions.

9   The computed distributions were constructed by assuming that the albedo distribution
10 for each distinct surface is represented by a normal distribution N($\bar{x}$,s), with $\bar{x} = \overline{\alpha_{ASD}}$
11 representing surface type and standard deviation (estimated from ASD measurements), s,
12 different for each surface type. Four distributions were constructed: clean ice N(0.56, 0.07),
13 dirty ice N(0.19, 0.05), shallow streams N(0.23, 0.09), and cryoconite holes N(0.10, 0.05).
14 Relative surface coverage of these four dominant surface types was derived at five distinct
15 time periods (1 June, 19 June, 18 July, 28 July, and 5 August) over the 2012 melt season from
16 Chandler et al. (2014; see Fig. 6a-g) to represent transient ice surface conditions, classified
17 here as "early summer ice", "dirty ice exposure", "melt", "darkening ice", and "late summer
18 ice", respectively (Table 3). A composite distribution for each distinct time step was
19 calculated as the weighted mean of surface type distributions, where the weights were
20 determined by their relative surface coverage area. Since Chandler et al. (2014) data are from
21 2012, results were not directly comparable with 2013 MOD10A1 data, but should capture
22 melt season evolution.

23   To compare with the computed distributions, high-quality 2012 and 2013 MOD10A1
24 data were used to construct observed albedo distributions at three spatial extents (50x50,
25 100x100, and 150x150 pixel extents; Fig. 1). The spatial resolution of the original MOD10A1
26 data is 463 m at nadir (exact resolution varies with overpass time), corresponding to study
27 areas of 23.2, 46.3, and 69.5 km^2 for the three spatial extents, respectively. Using a kernel
28 smoothing density estimator, the average probability density distribution was computed at
29 0.01 albedo bin widths (range from 0.05 to 1). The seasonal average albedo distribution was
30 calculated at the three spatial extents, and five-day average albedo distributions and spatial
31 averages were calculated for the 100x100 pixel scale for 2012 and 2013 MOD10A1 data.

3.6 Identification of snowfall events

To identify possible snowfall events in our study area and MODIS spatial extents, hourly precipitation and air temperature measurements collected by a meteorological station, hereafter 660 Met Station, installed near the ice sheet edge at the proglacial and ice sheet margin interface (Fig. 1), were examined. Near surface air temperature measurements from the shorter Base Met Stations time series (available from 8 – 26 June 2013) were also examined to estimate temperature differences between the proglacial and ice surfaces. Tundra near surface air temperature < 1 °C and precipitation > 0 °C were used as criteria to identify dates of likely snowfall events. To validate that solid precipitation fell, NASA's WorldViewer (https://earthdata.nasa.gov/labs/worldview/) was utilized to browse daily MODIS reflectance imagery (bands 7-2-1 and 3-6-7) to identify textural and brightness changes related to precipitation events.

3.7 Computation of relative melt rates

To examine seasonal changes in MODIS albedo, and estimate the importance of distinct surface types, relative surface melt rates were computed using the net shortwave solar radiation equation, observed values of incoming solar radiation from the Base Met Station on 16, 19, and 25 June, and broadband albedo values for computed and observed distribution methods. The observed incoming solar radiation values were averaged together and kept constant in the relative melt rate calculations to isolate the effects of albedo changes on melt. Net solar radiation (E_R) varies as a function of incoming solar radiation ($E_S^{\downarrow}$) and albedo (α_s), where units of energy are represented as W m^{-2}:

$$E_R = E_S^{\downarrow}(1 - \alpha_s) \tag{1}$$

Melt rate, defined as the heat needed to melt snow/ice when near-surface temperatures are ≥ 0 °C, was computed in units of m s^{-1} (Cuffey and Patterson, 2010):

$$M = E_R\left(L_f * \rho_w\right)^{-1} \tag{2}$$

where L_f is latent heat of fusion (3.34 x 10^5 J kg^{-1}); and ρ_w is density of water (1000 kg m^{-3}). Since the meteorological station datasets lack surface energy balance terms (i.e., net longwave

radiation, sensible and latent heat fluxes) required to compute the entire energy budget, calculating absolute melt rates was not possible. Instead, the percent difference in estimated melt rates was computed for each distribution relative to the early melt season ablation rates (mean of $4.40*10^{-7}$ m s^{-1} for "early summer ice" computed distribution; mean of $2.70*10^{-7}$ m s^{-1} for 31 May – 4 June observed MODIS distribution).

4   Results

4.1   Spatiotemporal patterns in ablation area albedo

Spatial variability of broadband α_{ASD} along the transect follows a consistent pattern on all three dates, averaging low values (0.50 ± 0.04) the first ~300 m, followed by increased albedo, reaching a plateau of 0.64 ± 0.07 at ~600 m, and remaining nearly constant with the exception of a dip to 0.44 ± 0.02 at ~900 m (Fig. 2a). While discrete α_{ASD} observations often differ from the nearest observation made at another transect time due to slight day-to-day changes in sample location (Fig. 2a), data averaged in 50 m bins covary spatially along the transect gradient (Fig. 2b). The spatial variability of broadband α_{ASD} is considerable and varies between a minimum of 0.14 (19 June) and a maximum of 0.75 (16 June; Table 1). The high variability in discrete α_{ASD} values over short distances (Fig. 2a) is indicative of the heterogeneous surface that characterizes the field site and surrounding ablation area, not necessarily captured in α_{base} observations.

Temporal variability in daily average α_{base} follows a non-linear decline from 8-26 June 2013 starting at 0.49 and ending at 0.34 (Fig. 3). An increase in α_{base} of 0.11 between June 12 and 16 might be related to tilt errors, which influenced what part of the increasingly heterogeneous surface the instruments were monitoring. Indeed, the net lowering of α_{base} by 0.15 between 8-26 June is consistent with $\alpha_{MOD\ Pixel\ 1}$ and $\alpha_{MOD\ Pixel\ 2}$ observations from June to mid-August. $\alpha_{MOD\ Pixel\ 1}$ and $\alpha_{MOD\ Pixel\ 2}$ drop from values slightly above 0.5 in June to 0.24 and 0.37, respectively, around mid-August. In between these dates, sudden increases in albedo could be caused by occasional snowfall events (Fig. 3). Potential snowfall events were identified as time periods with 660 Met Station temperatures < 3 °C that coincided with precipitation events. The 3 °C offset means that elevations 470 m higher than the 660 Met Station would be at freezing, given an environmental lapse rate of 6.4 °C km^{-1}, which is about

the elevation where the dark band region begins. By using this method to identify snowfall events, a brief event likely occurred on 28-29 June (Fig. 3) raising MOD10A1 albedos from 0.31 to 0.53 between 27 June and 30 June, respectively. July MOD10A1 albedos exhibited some temporal variability, but were generally lower at the end than the start of the month. It is unclear if they were triggered by snowfall events. While precipitation events occurred several times on the tundra in July, it is unknown if these events extended to the ice sheet and if temperatures were sufficiently cold to trigger snow rather than rain. August MOD10A1 albedo increased from early to late in the month with a snowfall event on ~18 August, triggering large increases in albedo to values above 0.75. High-quality daily average broadband $\alpha_{ASD\ Pixel\ 1}$ and $\alpha_{ASD\ Pixel\ 2}$ data don't exhibit the slight increase in α_{base} at the end of June (0.04 from 22-26 June), which may be reflected by differences in footprint sizes, a lower α_{ASD} sampling frequency, and α_{base} tilt errors. Instead, $\alpha_{ASD\ Pixel\ 1}$ and $\alpha_{ASD\ Pixel\ 2}$ data exhibit a steady decline over the month of June, while $\alpha_{MOD\ Pixel\ 1}$ and $\alpha_{MOD\ Pixel\ 2}$ data remain relatively constant over the same time period. Absolute magnitudes among the three ground- and satellite-derived albedo products diverge due to sensor, wavelength range and spatial resolution differences. However, all products have higher albedo values at the beginning as compared with the end of the month of June, prior to the 28-29 June snowfall event.

Albedos of dirty and clean ice surfaces are distinctly different for each ablation stake site (Table 2). Broadband α_{ASD} spectra made within 30 m of ablation stakes were individually assessed to classify each surface type into two distinct groupings: clean and dirty ice. Only Site D had both dirty and clean ice surfaces. Manual inspection of individual spectra at Site D confirm that samples with $\alpha_{ASD} < 0.4$ are qualitatively similar to typical spectra for wet or debris rich ice as shown in Pope and Reese (2014), and distinctly different from values of $\alpha_{ASD} > 0.4$.

4.2   Melt season albedo distributions

4.2.1   2013 computed and observed distributions

Computed albedo frequencies using typical albedo values for four distinct surface types (Table 3) and changing area fractions of these surfaces identified at a nearby site by Chandler et al. (2014) reveal a bimodal distribution as the melt season progresses (Fig. 4). The relative

1   strength of the first and secondary modes change as the fractional area of darker surfaces

2   expands from "dirty ice exposure" to "melt" distributions and onwards. At the start of the

3   melt season, the abundance of lighter surfaces coincides with a higher probability of high

4   broadband α_{ASD} values. Here, snow and clean ice surfaces dominate and gradually degrade,

5   exposing the impurity-rich surface underneath. As darker surfaces progressively populate the

6   ablation area with the onset of the melt season, computed albedo distributions predict a

7   concomitant higher probability of lower albedo. Thus, there is an apparent dichotomy

8   between darker and lighter surfaces 'competing' to control the overall albedo distribution of

9   the ablation area. A transition towards a distribution biased towards lower albedo values is

10   due to darker surfaces shifting the overall distribution, and is consistent with high-quality

11   broadband α_{ASD} distributions (Fig. 5). Relative melt rates increase sharply (by 25.7%) from

12   "dirty ice exposure" to "melt", coinciding with a strengthening of the second, lower mode in

13   the computed albedo distribution (Table 4). Once the secondary mode is established, a smaller

14   increase in melt rates occurred as the mode strengthens between "melt" to "darkening ice"

15   and finally to "late summer ice" (6.7% and 9.1%, respectively).

16   Observed 2013 MOD10A1 albedo distributions at three spatial extents (Fig. 6) reveal

17   that the bimodal distributions (cf. Fig. 4) are manifested in reality at the 100x100 MODIS

18   pixel (px) extent (i.e. 46.3 km^2). While the spatial extent of the MOD10A1 sample influences

19   the seasonal average albedo distribution, two distinct surface types - dark and light surfaces -

20   dominate the seasonal signal (Fig. 6). At the smallest spatial extent (50x50 px – i.e., 23.2

21   km^2), lower albedos from darker surfaces of the lower ablation area control the density

22   distribution, while at the largest spatial extent (150x150 px – i.e., 69.5 km^2), the probability

23   distribution is primarily influenced by higher albedos from lighter surfaces (e.g., snow) of the

24   upper ablation area. The central tendencies of each mode are ~0.46 and ~0.72, which are

25   much larger than in the computed distributions (~0.18 and ~0.56; cf. Fig. 4).

26   The bimodal distribution identified in the observed 100x100 px MODIS albedo

27   distribution in 2013 (Fig. 6) is the result of snow and ice surfaces characterizing the two

28   peaks, as each mode centers around typical values of snow and clean ice, respectively. As

29   such, the observed MODIS bimodal distribution is associated with a transition from ice to

30   snow, rather than a change from clean to dirty ice, which caused the two modes in the

31   computed distribution (Fig. 4). Indeed, analysis of 2013 meteorological observations reveal

32   that short term snowfall events that fell on top of the underlying ice can result in variations in

13

1 ablation area albedo (Fig. 3). In 2013, the bimodal distribution at the 100x100 px spatial

2 extent is likely the result of snow deposition or redistribution of blowing snow (e.g., Gorter et

3 al., 2014; Lenaerts et al., 2014) on top of the ice surface (Fig. 6).

4 MOD10A1 albedo at the 100x100 px (i.e., 46.3 km^2) spatial extent transitions from a

5 unimodal distribution with high albedo values at the start of the melt season (31 May – 4

6 June), to a bimodal-like distribution with intermediate albedo values at mid-melt season (20 –

7 24 June), and shifts abruptly to a new, unimodal distribution with low albedo values at peak

8 melt season (30 July – 3 August; Fig. 7). By assuming an unchanged radiation budget, the

9 relative impact of albedo distribution changes on melt rates was quantified. The abrupt shift

10 from a lighter- (high albedo) to darker-dominated (low albedo) surface corresponds to an

11 observed melt rate percent difference increase of 51.5% between 10 – 14 July and 20 – 24

12 July five-day average albedo distributions (Fig. 8). Before and after this shift, melt rates

13 changed much less from each five-day average, ranging between ~10-30%, with the exception

14 of the dramatic drop of 103.3% when the melt season ends in late August.

15 The bimodality seen in the 30 June – 4 July five-day average distribution (Fig. 7)

16 coincides with a brief period of higher MODIS albedo values (~0.6 – 0.7), indicative of snow.

17 Identification of a snowfall event on 28-29 June 2013 (Fig. 3), confirms the source of the

18 bimodal distribution observed in the 30 June – 4 July five-day average (Fig. 7), corresponding

19 to a brief "jump" in the probability density distribution to higher albedos.

20

21 4.2.2 Differences between 2012 vs. 2013 observed albedo distributions

22 While the 2013 MODIS albedo bimodal distribution shown in Fig. 6 and 7 are a result of

23 snow and ice albedo, analysis of MODIS 2012 data reveal a more complex, multi-modal

24 albedo distribution (Fig. 9). These distributions cannot be explained by the presence or

25 absence of snow and ice alone. The 2012 MODIS observations are characterized by generally

26 lower albedo, with six out of nine five-day average albedo distributions ranging mostly

27 between 0.2 and 0.5, compared to three out of nine five-day average albedo distributions in

28 2013 (cf. Fig. 7 and 10). These low albedos are confirmed by the average seasonal MODIS

29 2012 albedo distributions, where a higher probability of albedos are centered on ~0.35,

30 compared to two peaks at ~0.45 and ~0.7, in 2013 at the 100x100 px spatial extent (cf. Fig 6

31 and 9). The higher probability of these very low albedos observed in 2012 are likely due to

dust, sediment, and impurity-rich ice in the so-called 'dark-band' region (Wientjes and Oerlemans, 2010). The identification of this dark zone feature is presented in section 4.2.3.

4.2.3   2012 vs. 2013 spatial maps

The presence of the dark band region is confirmed by the diagonal band of very low albedo (< ~0.35) in the 2012 MODIS seasonal average at the 100x100 px extent (Fig. 11). However, the presence of the dark band region is not visible in 2013, where albedo gradually increases from west to east (Fig. 12). The lack of the dark zone feature in 2013 is likely due to snow covering the dark band for most of the season. Overall, 2012 exhibits substantially lower ablation area albedos (Fig. 10), while 2013 reveals higher ablation area albedos in the MODIS spatial averages (Fig. 7). The large inter- and intra-annual variability in MODIS ablation area albedo may be indicative of the large spatial variability in surface types that characterize the lower elevations of the ablation area. Alternatively, a larger distribution in cryoconite hole coverage may have also contributed to low albedos (~0.25) observed in the 2012 MODIS seasonal averages (Fig. 9).

4.3   Relative melt rates

Observed ablation rates, derived from stake readings, are typically higher for dark surfaces (dirty ice and streams) than light surfaces (clean ice; Fig. 13). Clean ice surfaces have higher broadband α_{ASD} values (mean of 0.57), corresponding to lower average ablation rates ($5.38*10^{-7}$ m s^{-1}). In contrast, dirty ice and stream surfaces have lower mean broadband α_{ASD} values (0.24), corresponding to higher average ablation rates ($6.75*10^{-7}$ m s^{-1}). The observed mean difference between light and dark surface ablation rates is $1.37*10^{-7}$ m s^{-1}. Melt rate calculations (Eqn. 1 and 2) resulted in a lower average ablation rate for clean ice surfaces ($4.24*10^{-7}$ m s^{-1}) and a higher average ablation rate for dark ice surfaces ($7.56*10^{-7}$ m s^{-1}), corresponding to a mean difference of $3.33*10^{-7}$ m s^{-1}. Differences between observed and calculated melt rates could be due ablation stake measurement errors and simplification of calculations (e.g., no consideration of longwave radiation or turbulent heat fluxes). Regardless, in both cases relative melt rates between light and dark surfaces are considerably different, and thus useful for investigating seasonal melt rate changes as described below.

The spread in observed clean ice broadband albedo values results in greater variability in observed ablation rate estimates (Fig. 13). In contrast, minimal broadband albedo variability is observed for dirty ice surfaces. Few dirty ice albedo measurements were sampled as compared to clean ice surfaces. Differences in observed ablation rates for streams are due to a lack of albedo measurements taken over these surfaces. While ablation rates were measured at several ablation stake stream sites, only occasional α_{ASD} measurements were collected over these surfaces. Considerable spread in ablation rates for stream observations could be explained by varying stream depth (Legleiter et al., 2014). The depth of these ice streams determines the attenuation and scattering of radiant energy, thereby influencing the observed albedo measurements. Sensible heat flux from the stream water, not accounted for in radiative estimates, may also be a mechanism for increased melting.

5 Discussion

5.1 The importance of surface types in observed and computed ablation area albedo

GrIS ablation area albedos are strongly influenced by the presence or absence of impurity-rich debris on its surface. Clean ice and dust-covered, dirty ice have distinctly different albedos, resulting in a left-skewed albedo distribution at mid- and end of June (Fig. 5). This pattern is supported by computed and remotely-sensed albedo distributions, revealing that a multimodal distribution develops seasonally. A modest melt or snowfall event can trigger a sudden switch from a high to low albedo mode or vice versa, drastically changing ablation rates. These findings suggest that shifts in dominant surface type from snow to bare ice, and clean ice to impurity-rich surfaces are an important driver in abruptly increasing seasonal ice sheet melt rates.

The first quality-controlled in situ ablation area albedo dataset collected along a 1.25 km transect during three days in June 2013 is presented. Albedo data collected during in situ transect dates resemble an early summer ice surface classified in Chandler et al. (2014) and Knap and Oerlemans (1996; Fig. 4). Here, remaining snow cover and superimposed ice gradually melts, revealing underlying impurities and cryoconite holes. Visual assessment and continuous monitoring in the field revealed that the ice surface along the transect was snow-free from 8-26 June 2013. This period corresponds to a non-linear decrease in albedo (Fig. 3).

Accumulation of exposed below-surface impurities (Wientjes and Oerlemans, 2010), the gradual erosion of snow patches in local depressions on the ice surface (van den Broeke et al., 2011), as well as the activation and development of the hydrologic system and cryoconite hole coverage (Chandler et al., 2014) may mitigate the rate of change in ablation area albedo. Turbulent sensible heat fluxes from adjacent pro-glacial areas provide an additional explanation for the non-linear decline in ground albedo measurements, serving to limit the melt-albedo feedback's influence (van den Broeke et al., 2011).

Under the assumptions that distinct surface types follow a normal distribution, a bimodal probability distribution preferentially develops as ablation area albedo decreases rapidly over the melt season due to development of an efficient meltwater drainage system, increase in cryoconite hole coverage, and accumulation of debris-rich sediments (Fig. 4). An increase in debris-rich and stream surfaces over the melting season (Fig. 4) is likely responsible for the enhanced frequency of low albedo values identified in the observed α_{ASD} distribution from 16-25 June (Fig. 5). However, the observed changes at transect sites appear to be more gradual than for the MODIS data (Fig. 7 and 10). This may be due to a lack of snow cover influencing the local albedo distribution and a lower temporal sampling frequency. The lack of a pronounced secondary mode with lower albedo values in the observed left-skewed distributions (Fig. 5) compared to the modeled bimodal distribution (Fig. 4) may be related to different melt season conditions (2012 vs. 2013), and corresponding range of surface types captured along the transect, which undersamples dark surfaces (e.g., dirty ice and stream surfaces; Table 2). While, Chandler et al. (2014) surface types cover a wider range of surface types, and thus, albedos.

Compared to reality, the computed distribution (Fig. 4) probably overemphasizes each mode and does not account for darkening due to ice crystal growth over the melting season. The observed albedo distributions reveal abrupt and variable shifts in the seasonal albedo distribution (Fig. 7 and 10). At certain spatial extents, these albedo distributions transition from a high- to low-dominated mode (Fig. 6), enabling enhanced melt rates (Table 4 and Fig. 8). Alexander et al. (2014) also observed bimodal albedo distributions for Greenland's ablation area by analyzing MAR and MODIS products between 2000-2013. Alexander et. al. (2014) attribute the dominant modes to the presence of snow and ice (and firn). This is in agreement with the analysis of the 2013 conditions, but disagrees with 2012 conditions. This discrepancy could be due to the larger study area that includes areas unaffected by dust from

1 deposition and outcropped ice layers, and a thirteen-year averaging period suppressing outlier

2 years like 2012 used in Alexander et al. (2014).

3   The bimodal albedo distribution and shift from a higher to a lower albedo mode

4 centering at values below 0.4 (Fig. 4, 7, and 10) indicate that a switch in dominant surface

5 type (i.e., from light to dark) during the melt season, and not solely grain size metamorphism,

6 are largely responsible for lowering albedo in snow-free ablation areas. Furthermore, results

7 from the MODIS data (Fig. 7 and 10) suggest that a transition from a light- to dark-dominated

8 surface is abrupt rather than gradual, likely associated with the addition and removal of snow.

9 The transition is more gradual in the left-skewed observed (Fig. 5) and computed albedo

10 distributions (Fig. 4), likely reflecting changes in impurity content and different time stamps.

11 Consistent with Chandler et al. (2014), the initial drop in MODIS ablation area albedo is

12 likely due to the transition from dry to wet, and patchy snow surfaces. Successive lowering of

13 albedo after snow melt is predominantly due to an increase ice crystal size (Box et al., 2012)

14 and possibly also by expansion of darker surface area coverage (e.g., cryoconite holes,

15 accumulation of impurities, and stream organization) and melting of dust-enriched ice layers.

16 These distributions correspond to percent differences (e.g., 51.5% between 10 – 14 July and

17 20 – 24 July five-day averages) in melt rate estimates that are substantial over the melt season

18 (Table 4 and Fig. 8), and highlight the importance of considering the albedo of ablation area

19 surface types. The higher melt rates associated with darker surfaces (Fig. 13) may lead to

20 lighter surfaces becoming topographically prominent. In theory, this should enhance sensible

21 heat transfer to the lighter surfaces, increasing their ablation. Future studies should consider

22 quantifying the effects of surface roughness on ablation area albedo (e.g., Warren et al., 1998;

23 Zhuravleva and Kokhanovsky, 2011), and the possibility of enhanced ablation of light

24 surfaces following upon adjacent, dark surface ablation.

25   Recent studies have proposed scenarios of future atmospheric warming, where excess

26 deposition of light-absorbing impurities (Dumont et al., 2014) and black carbon from

27 increased forest fire frequency or incomplete fuel combustion (Keegan et al., 2014), will

28 promote accumulation of impurities, contributing to amplified surface melting. If these

29 findings turn out to be true, these effects will likely be exacerbated in southwest Greenland's

30 ablation area, where continued negative albedo trends (Stroeve et al., 2013), and increasingly

31 warmer average summer temperatures (Keegan et al., 2014), in conjunction with bare ice,

32 light-absorbing impurities, and cryoconite holes, are expected to dominate.

5.2   Insights from 2012 and 2013 melt seasons' albedo distributions

The spatial distribution of snow cover and background bare ice albedo is important for understanding temporal changes in 2012 and 2013 MODIS albedo distributions (Fig. 11 and 12). Compared to 2013, snow melt in 2012 was more pronounced and reached higher elevations (Tedesco et al., 2014), allowing the dark band feature to be exposed, resulting in a lower seasonal albedo mode (Fig. 9).

The large albedo distribution changes from one MODIS five-day average to another in 2012 (Fig. 9) is likely due to variability in meltwater ponding on the ice surface, and perhaps deposition of wind-blown dust from tundra regions, and not necessarily increases in melted-out debris from internal ice layers at such short timescales. However, exposure of dust and sediment-rich ice surfaces probably caused the high probability of considerably low 2012 MODIS albedo values relative to 2013. This is expected since it was identified as an extreme melt year with early onset snow melt (e.g., Nghiem et al., 2012; Tedesco et al., 2013; Fig. 9 and 10), while 2013 was a normal melt-year in the 1979-2013 context (Tedesco et al., 2014). Given the coarse resolution of the MODIS pixel, it is likely that it averages out finer scale details of distinct surface types (e.g., dirty ice and cryoconite hole surfaces) along the ice sheet edge. It is hypothesized that higher spatial resolution satellite imagery may be able to capture such regions closer to the ice sheet margin. We postulate that the area of these regions may grow in size over the melting season as demonstrated on local scales by Chandler et al. (2014) in situ observations.

The bimodal distribution observed in the 2013 MODIS data (Fig. 4) appears to be governed by the relative extent of clean ice and snow surfaces. This aligns with findings from current SMB models, as the majority of variability in the overall Greenland ablation area albedo is driven by the deposition, change, and removal of snow (Alexander et al., 2014; Van Angelen et al., 2012). However, 2012 MODIS albedo distributions cannot be explained by transitions from snow to ice and vice versa. Instead, the 2012 MODIS albedo distributions likely reflect abrupt shifts in ablation area albedo from the exposure of impurities on the ice surface in the so-called "dark-band" region as well as ice crystal growth and expansion of dirty ice areas, even with the presence of a few snowfall events. As such, dust and impurities on Greenland's ice sheet surface can influence surface albedo in the ablation area. The current

state of SMB models are capable of simulating albedo as a function of meltwater ponding (Alexander et al., 2014) and impurities from atmospheric dust deposition on snow (Van Angelen et al., 2012). The models might be improved by incorporating the melting out of dust and sediments in outcropped ice layers, found in the dark band region.

6   Conclusions

A first high-quality in situ spectral albedo dataset collected along a fixed transect is presented for southwest Greenland's ablation area. Previous studies have attributed an increase in melt season duration, less snowfall accumulation, enhanced snow grain metamorphism rates and ice-albedo feedback as primary mechanisms for lowering ablation area albedo. Here, we demonstrate an additional control on albedo in the ablation area, namely the distribution of distinct surface types such as snow, clean ice, impurity-rich ice, melt ponds and streams, and also examine their modulation on surface ablation. The spatial extent of each of these surface types result in a multi-modal albedo distributions in the ablation area. Analysis of MODIS data suggest that a multi-modal distribution and consequentially, a shift from light to dark-dominated surfaces, and sensitivity to melting of outcropped ice layers, characterize seasonal changes in Greenland's ablation area, and therefore, melt rates.

Continued atmospheric warming coinciding with a darkening ice surface will increase the ice sheet surface meltwater production and runoff. Here, we show the importance of the distribution of dirty ice surfaces, which are likely the result of accumulation of impurities melted out from internal ice layers (at longer timescales; e.g., summer 2012) rather than contemporary deposition of atmospherically transported dust (except perhaps at short time-scales). Future research should investigate the importance of surface accumulation of impurities and if its surface area can change to significantly influence GrIS albedo and surface ablation. Finer spatial resolution satellite imagery is needed to adequately characterize the high spatial variability in surface types and their corresponding albedo in the ablation area of the GrIS. Analysis of spatio-temporal variability in albedo using higher spatial resolution imagery may be needed to adequately characterize surface types, particularly for dust and sediment-rich surfaces, to improve our understanding of the contribution of ablation area albedo to GrIS mass loss.

Acknowledgements

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1   Table 1. Descriptive statistics for high-quality albedo transects. SZA and CC listed for Base Met Station only. "Brd" is used to abbreviate

2   broadband.

Transect Date	Start Time	End Time	Min Brd α_{ASD}	Max Brd α_{ASD}	Mean Brd α_{ASD}	Daily Average α_{base}	Daily Average α_{top}	Min SZA (°)	Max SZA (°)	Mean SZA (°)	Min CC	Max CC	Mean CC
16-Jun	10:32:33	11:53:57	0.260	0.754	0.550	0.404	0.636	45.615	50.454	47.828	0.135	0.176	0.157
19-Jun	10:39:30	11:35:59	0.141	0.730	0.532	0.316	0.541	46.449	49.925	48.093	0.045	0.084	0.065
25-Jun	10:20:29	11:11:00	0.210	0.670	0.490	0.333	0.525	47.963	51.525	49.677	0.119	0.138	0.125

1 Table 2. Average broadband α_{ASD} within a 30 m radius of ablation stake sites and classified by

2 surface type. The number within the parenthesis denotes the sample size.

Ablation stake sites	Clean surfaces	Dark surfaces
Site A	0.641 (4)	-
Site B	0.540 (4)	-
Site C	0.591 (7)	-
Site D	0.530 (4)	0.243 (2)
Site E	0.555 (5)	-

1   Table 3. Seasonal evolution (%) of four surface types at five distinct time steps approximated

2   from Chandler et al. (2014).

Time steps	Classified names	Clean ice	Dirty ice	Streams	Cryoconite holes
1 June	Early summer ice	100	0	0	0
19 June	Dirty ice exposure	90	3	1	6
18 July	Melt	60	20	1	19
28 July	Darkening ice	50	30	3	17
5 August	Late summer ice	40	40	6	14

3

1   Table 4. Percent difference in melt rate estimates for different albedo probability density

2   functions and averaged incoming solar radiation conditions at Base Met Station from 16, 19,

3   and 25 June, relative to 'early summer ice (1 June)' distribution.

Time steps	Classified names	Melt rate percent difference (%)
19 June	Dirty ice exposure	8.28
18 July	Melt	34.01
28 July	Darkening ice	40.73
5 August	Late summer ice	49.83

4
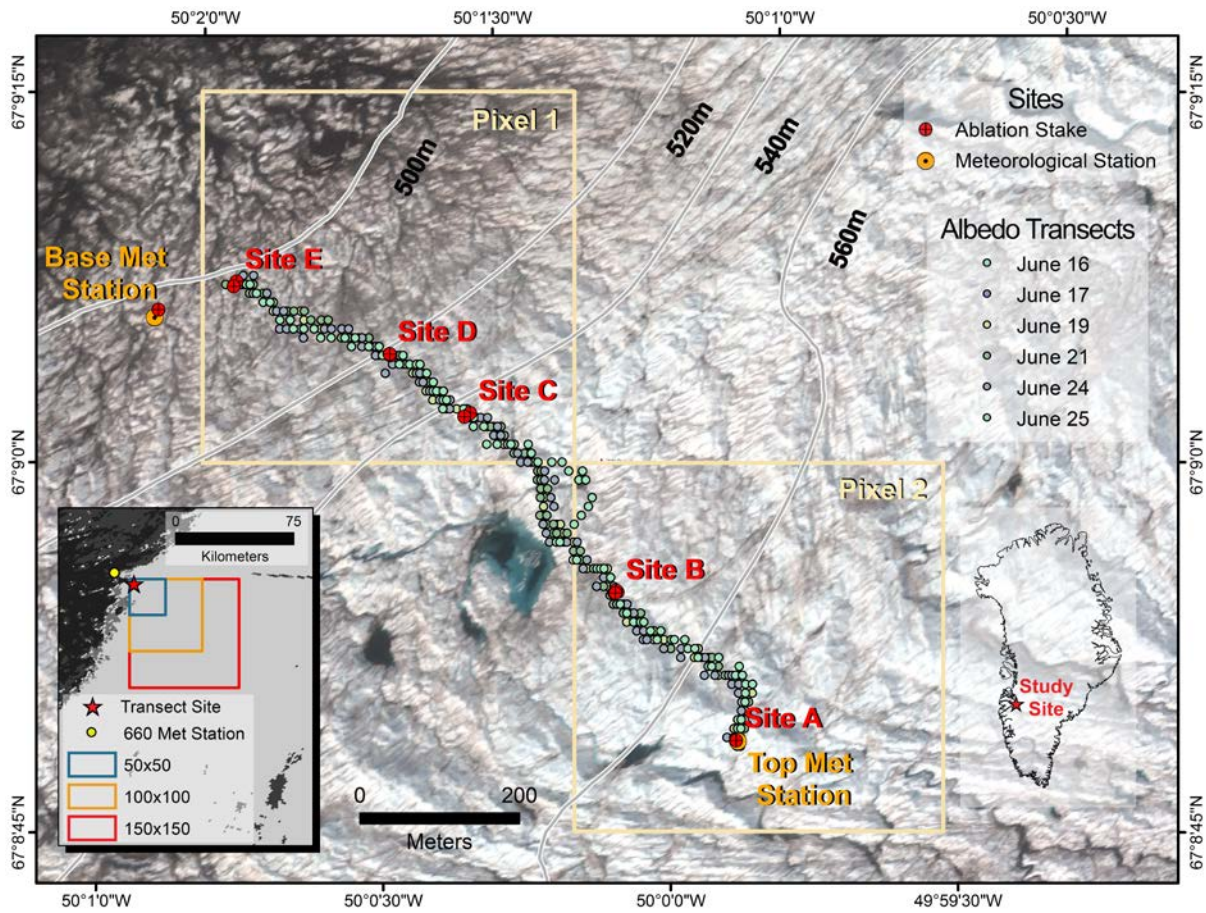
1
2   Figure 1. 23 June 2013 WorldView-2 true color image (bands 5, 3, and 2 RGB) of the study

3   site with elevation contours (m), MODIS pixel extents (yellow boxes), and location of the six

4   albedo transects, ablation stake, and meteorological station sites. Location of four MODIS

5   spatial extent regions overlaid on a 31 May 2013 MOD10A1 image (black box inset).
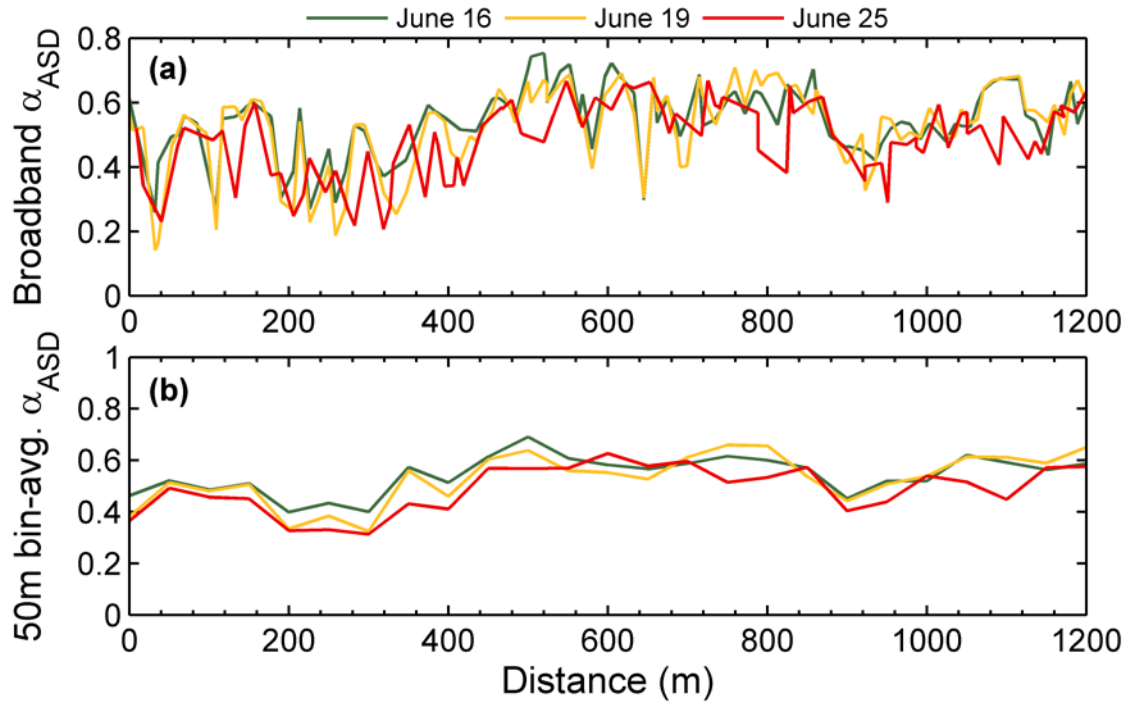
Figure 2. High-quality broadband α_{ASD} observations on 16, 19 and 25 June (a), and broadband α_{ASD} averaged in 50 m bins (b) along the length of the transect starting near Site E (0 m) and ending near Site A (1200 m).
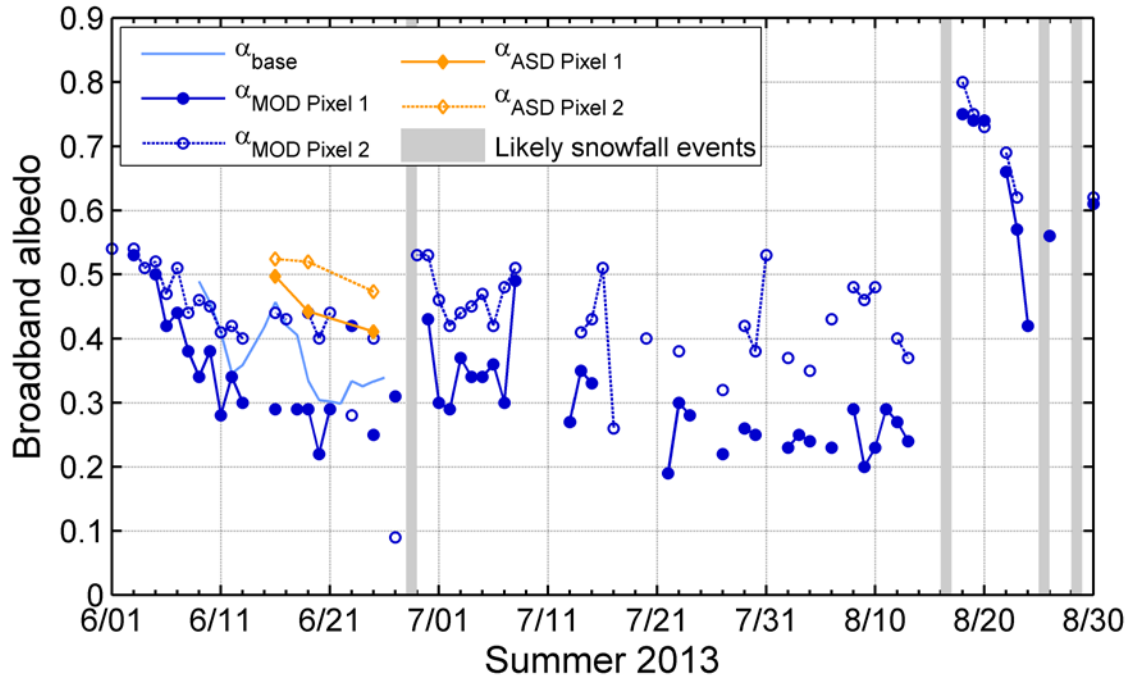
1
2 Figure 3. High-quality daily average broadband $\alpha_{ASD\ Pixel\ 1}$ and $\alpha_{ASD\ Pixel\ 2}$, α_{base} (for SZA <

3 70°), and $\alpha_{MOD\ Pixel\ 1}$ and $\alpha_{MOD\ Pixel\ 2}$ time series for the 2013 melt season. $\alpha_{ASD\ Pixel\ 1}$ and α_{ASD}

4 $_{Pixel\ 2}$ pixel-averaged values correspond to high-quality ASD transect dates 16, 19 and 25 June.

5 Grey shaded regions show likely snowfall events.

6
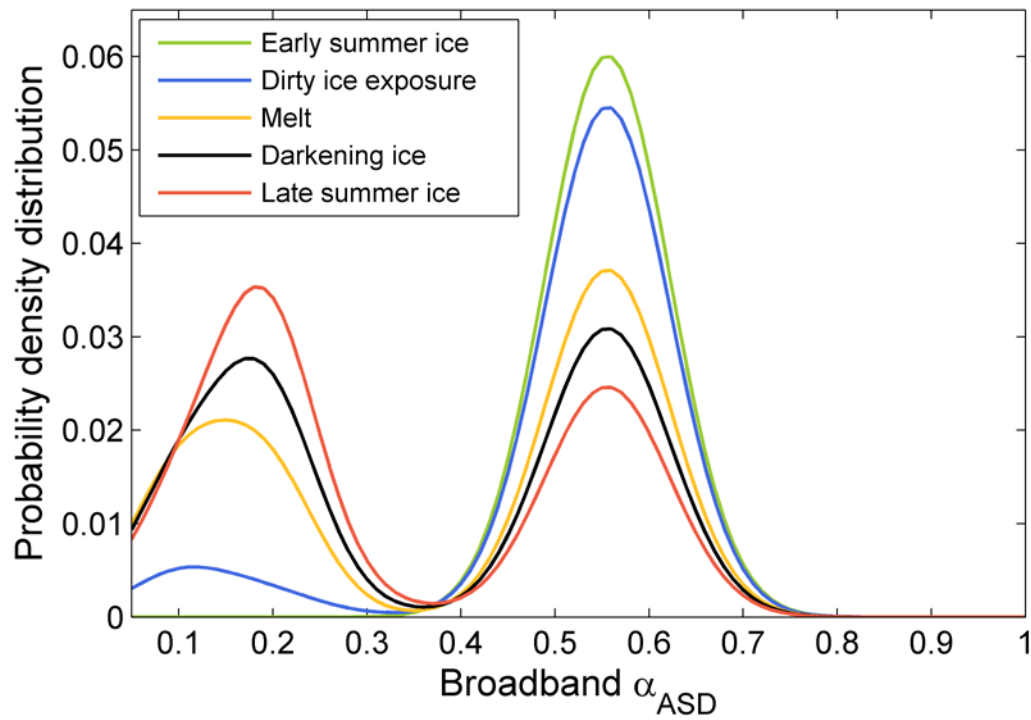
Figure 4. Computed albedo distribution for a nearby site of Chandler et al. (2014) simulated across the melt season based on observed broadband α_{ASD} values for dominant surface types, weighted by their relative surface area coverage. Each surface type is assumed to follow a normal distribution. Computed albedo distributions represent the sum of each surface type's probability distribution function.
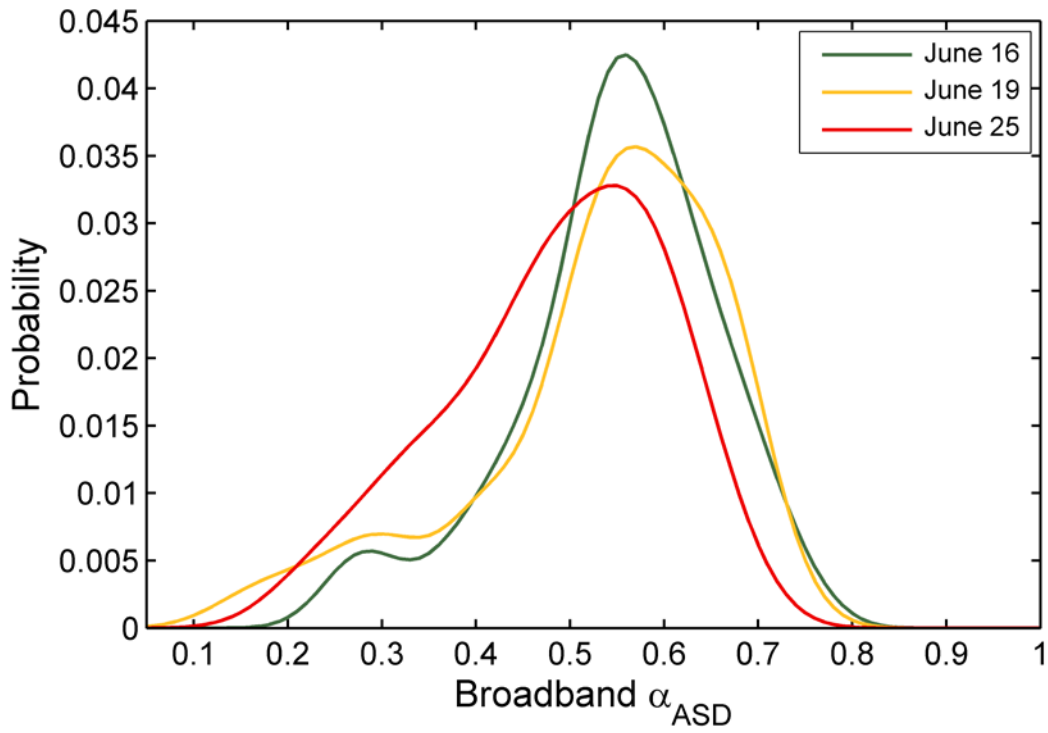
Figure 5. Observed distributions of high-quality broadband α_{ASD} transects on June 16, 19, and 25.
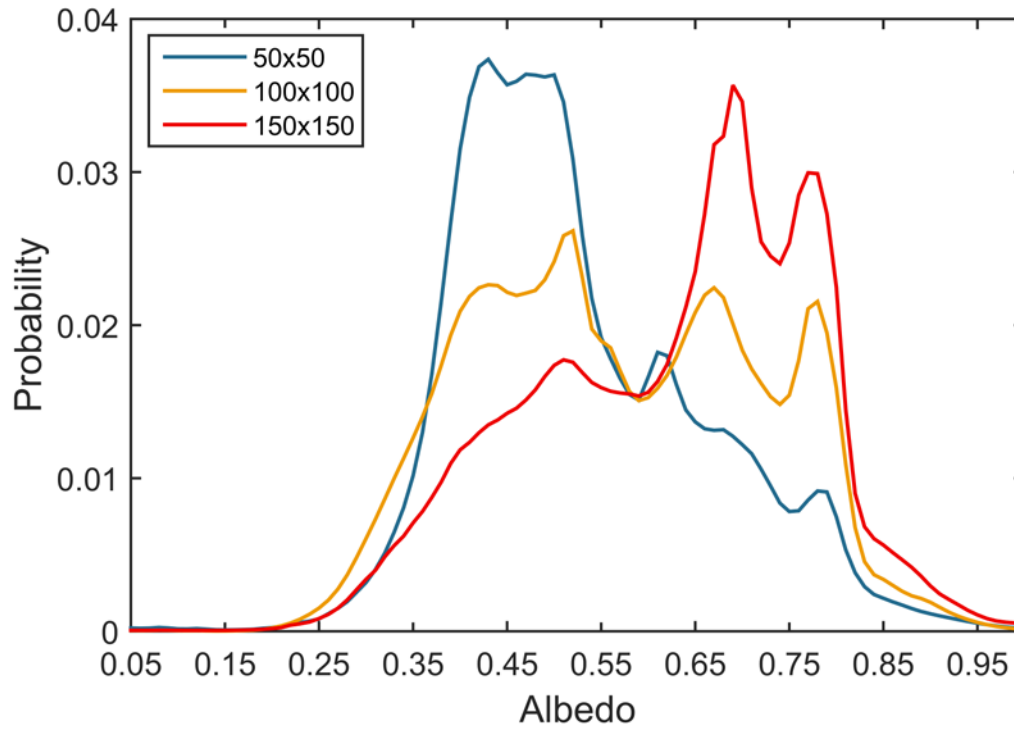
1
2 Figure 6. MOD10A1 2013 seasonal average albedo probability density distributions at three
3 spatial extents, 50x50 MODIS pixels (px), 100x100 px, and 150x150 px, respectively. The
4 bimodal distribution seen at the 100x100 px (46.3 km^2) spatial extent is likely the result of
5 almost equal area of snow and ice facies characterizing the two peaks. In contrast, the right
6 and left skew distributions of 50x50 px and 150x150 px illustrate the dominance of ice and
7 snow surfaces, respectively.
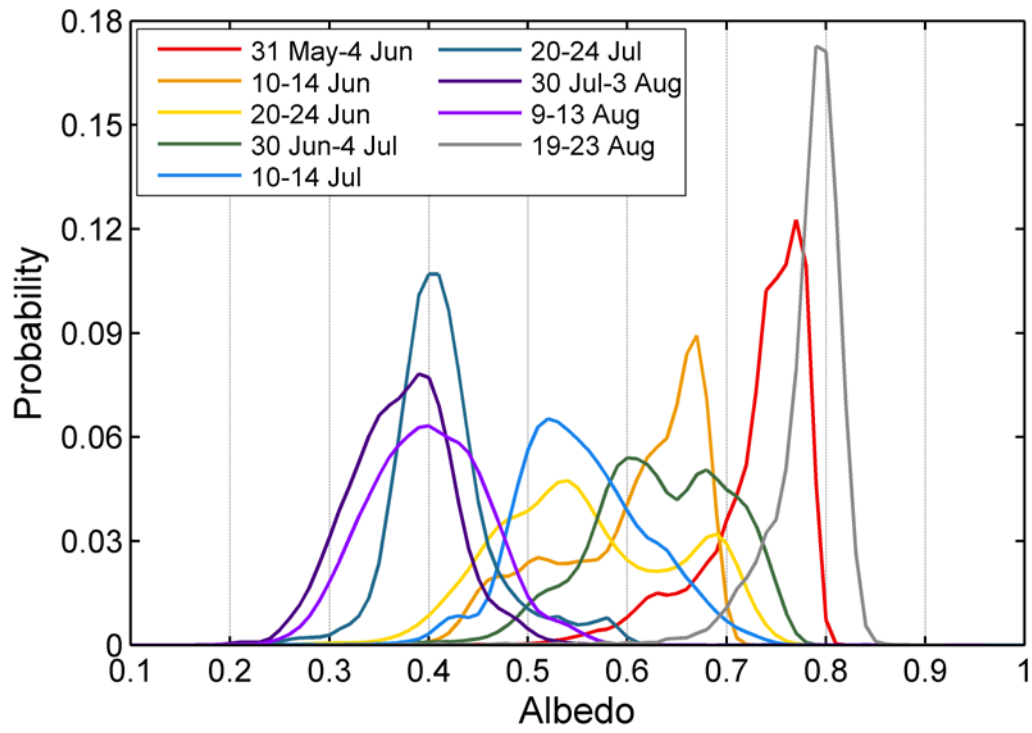
Figure 7. 100x100 px five-day averages over the 2013 melt season. Every other five-day average line is plotted.
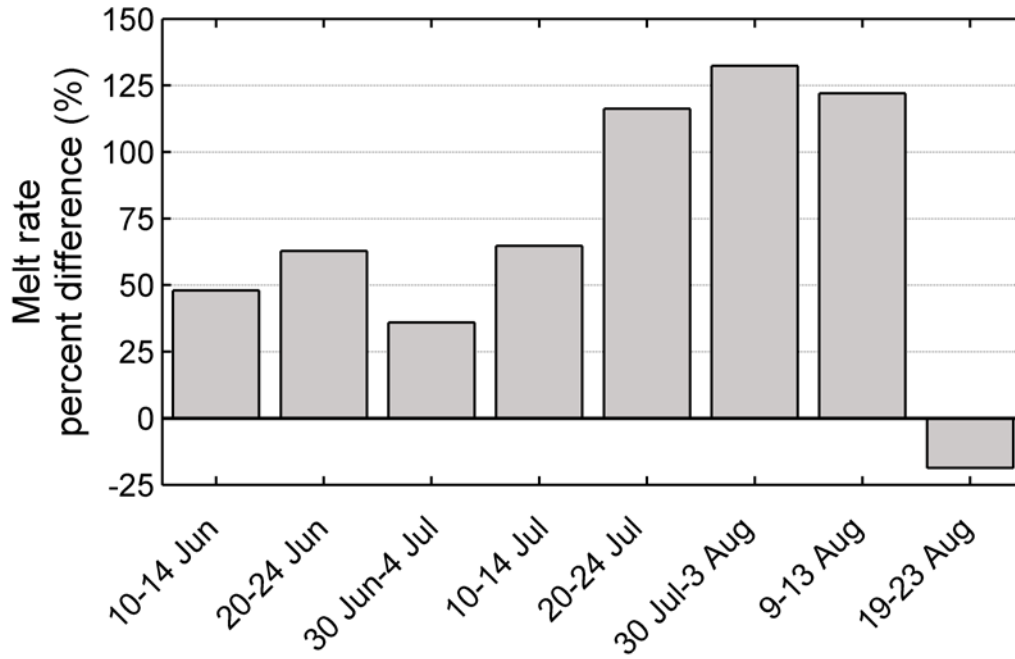
Figure 8. Percent difference in melt rate estimates for 100x100 px five-day average albedo distributions for the 2013 melt season, relative to 31 May – 4 June five-day average albedo distribution. Melt rates are calculated with identical radiation budget conditions to isolate the effect of albedo distribution changes.
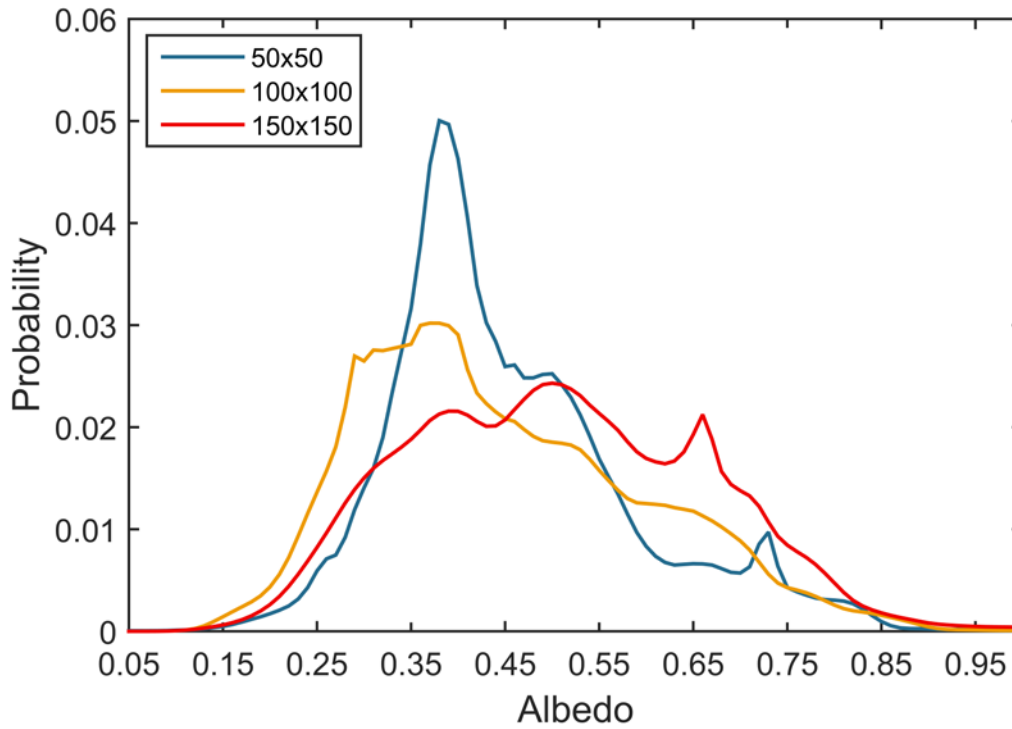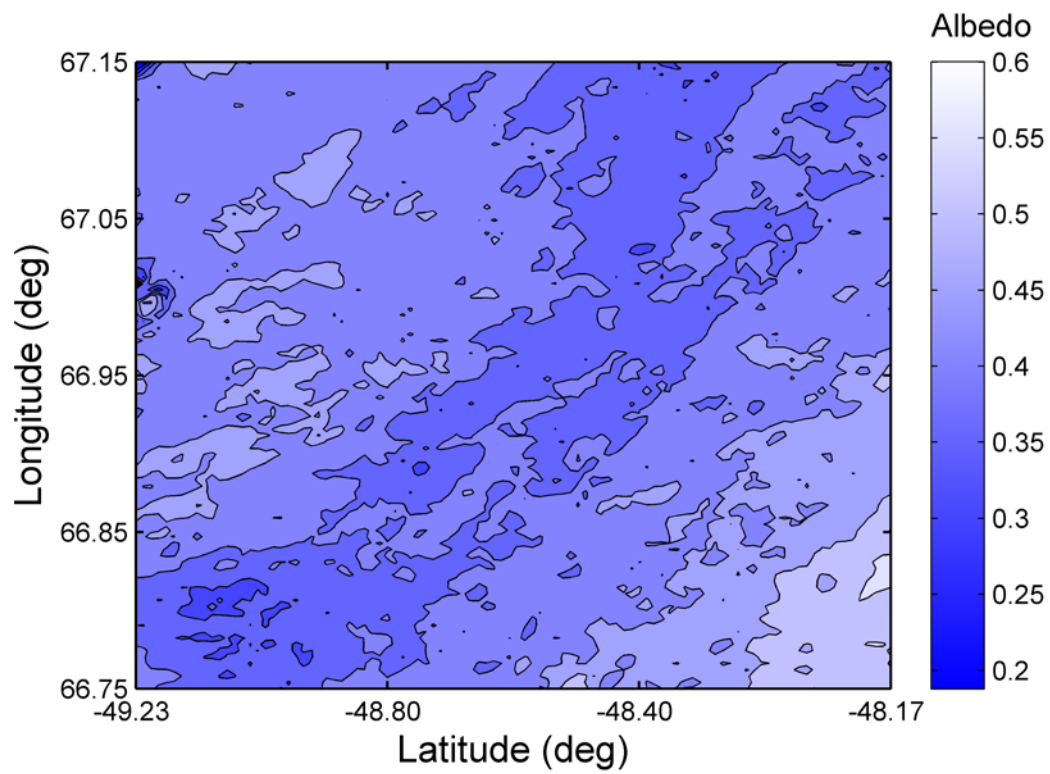
Figure 9. MODIS 2012 seasonal average albedo probability density distributions at three spatial extents. The MODIS 2012 seasonal average albedo probabilities for the 100x100 px and 150x150 px reveal a high probability of low albedo values (0.2 – 0.3). This is likely influenced by the expansion of the "dark band" region in these spatial extents.

1
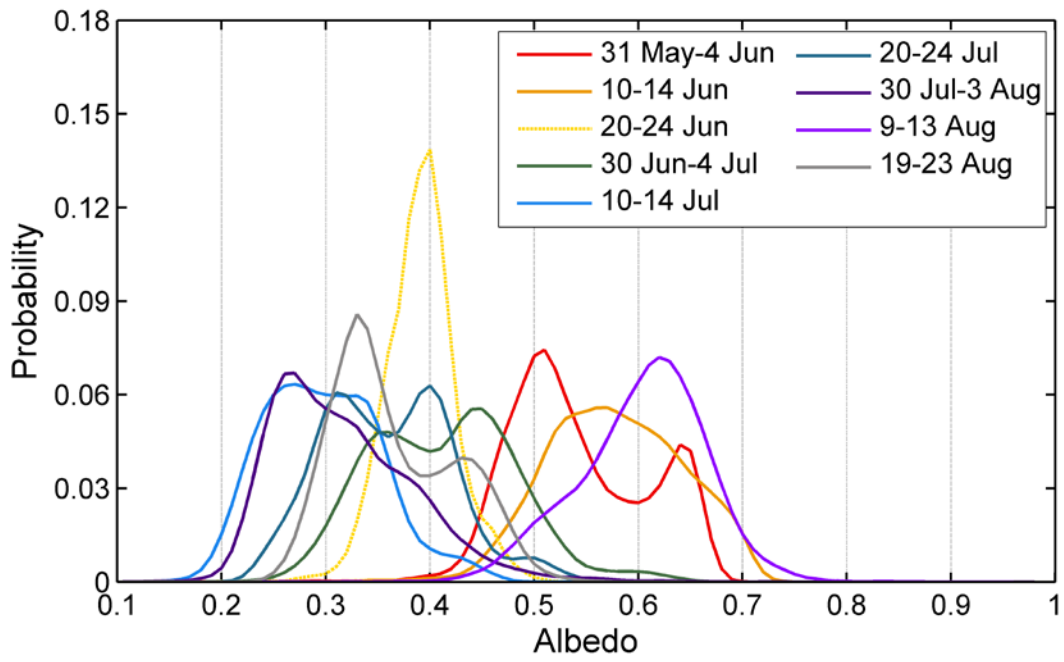
Figure 10. MODIS 100x100 px spatial extent five-day average albedo distributions for the 2012 melt season. Note, the 20 – 24 June five-day average (yellow stippled line) is most likely erroneous due to an outlier in the MODIS data on 21-22 June 2012.

Figure 11. MODIS 2012 seasonal average for the 100x100 spatial extent. A region of dark ice, known as the "dark band", extends through our study area (< ~0.35, shown in bright blue colors).
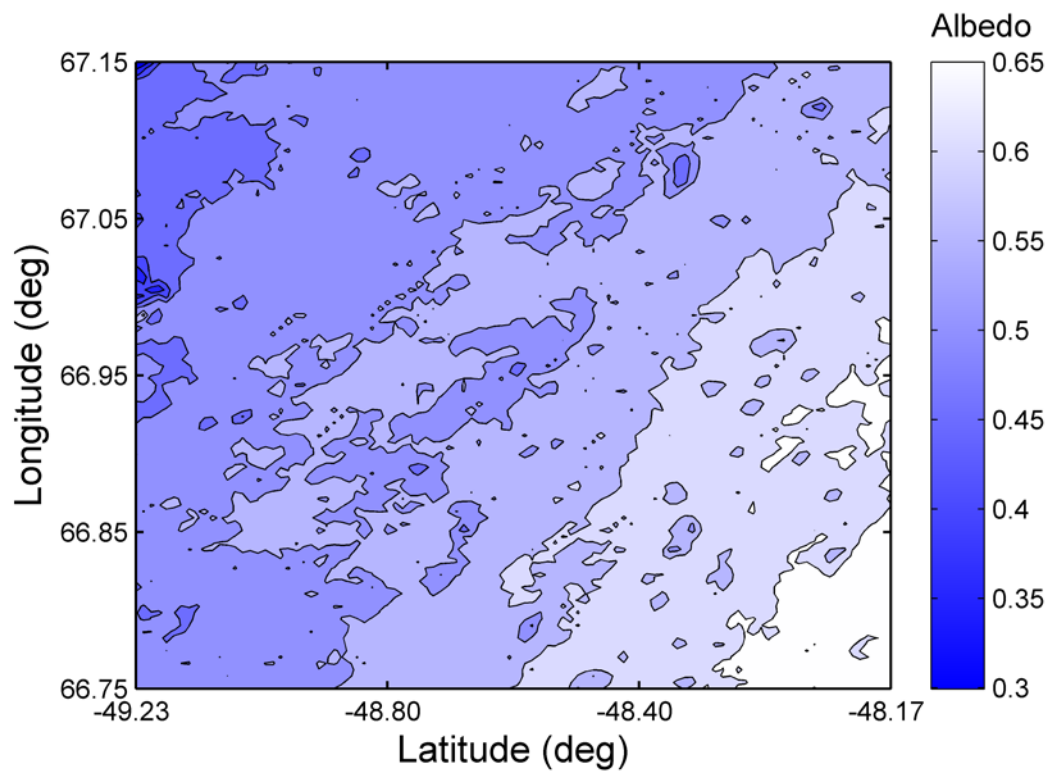
Figure 12. MODIS 2013 seasonal average for the 100x100 px spatial extent. Overall higher MODIS albedo values are observed in 2013, without a "dark band" region surface expression.
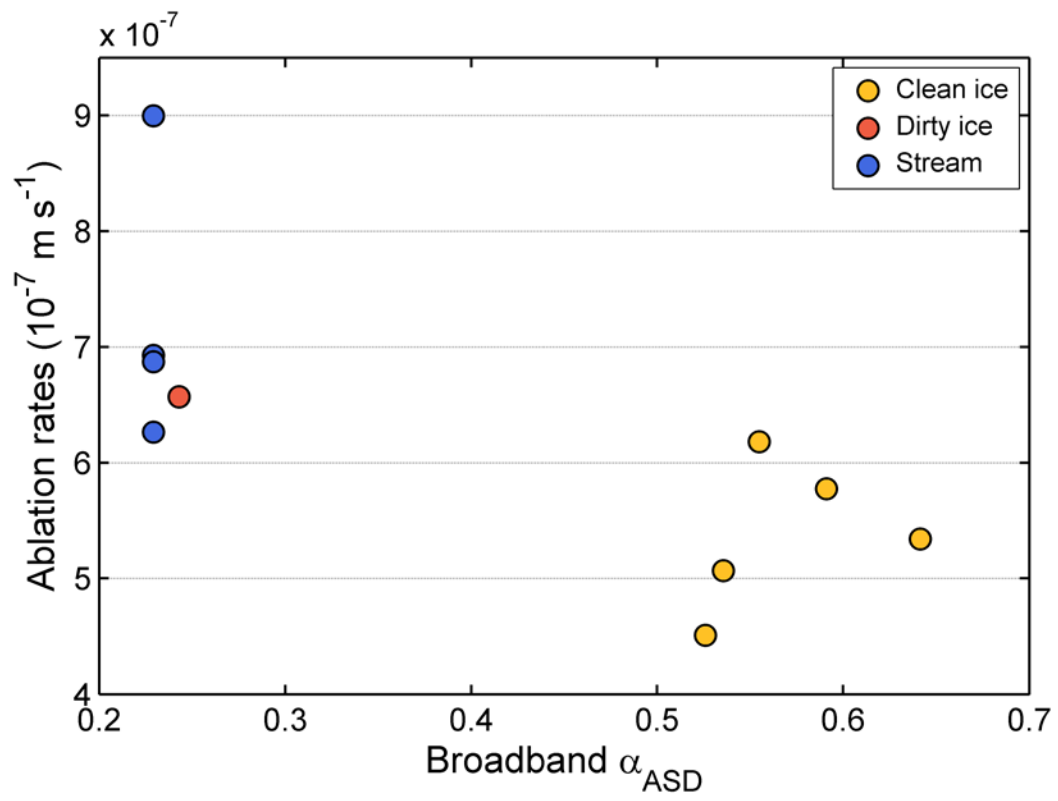
1
2   Figure 13. Observed ablation rates and broadband α_{ASD} for different ice surface types.

7   Appendix A

7.1   Field spectroscopy measurements

At the start of each transect, the ASD was calibrated to current hemispherical atmospheric conditions by orienting the RCR skyward, along a nadir-viewing angle. Subsequent measurements were taken with the ASD rotated 180° to view the ice surface. Under changing sky conditions, the instrument was recalibrated. Each transect consisted of ~100 sample locations, roughly 10 m apart. Despite changing ice conditions rapidly deteriorating temporary location markers, global positioning system (GPS) locations reveal that sample sites in consecutive transects were gathered in close proximity (Fig. A1). While samples were not taken from exactly the same sites, preventing a point-by-point comparison, transect sample distributions and smoothed spatial patterns can be analyzed for change over time. Sample sites along each transect were selected based on distance. If a spectrum site intersected with a stream, melt pond, or cryoconite hole, the nearest ice surface was sampled instead. To capture spectral albedo of different ice surface types, separate measurements of streams, dirty ice, and white ice were collected. At each sample location, five consecutive spectra consisting of 10 dark currents per scan and 10 white reference measurements were recorded and averaged.

Apparent outliers were identified using the Spectral Analysis and Management System software (SAMs) to identify outliers. Outliers were defined as physically unrealistic spectral albedo values (> 1.0) and raw spectra that were significantly different to the other spectra across the entire spectral range (visible and near-infrared wavelengths) taken for the same sample. For 16 June, 20 spectra were deemed outliers (total spectra collected = 555); 19 June, 17 spectra were deemed outliers (total spectra collected = 560); and 25 June, 12 spectra were deemed outliers (total spectra collected = 480). The outliers for these transect dates comprise less than 4% of all spectra collected, and thus, likely had an insignificant impact on the final albedo calculations. On 17 June, spectra with unrealistic > 1.0 values were collected, as will be shown in section 7.2. All data from this day were considered low-quality and removed from the dataset.

7.2 Quality-control of α_{ASD} data

To ensure a high-quality α_{ASD} dataset, an impact assessment of variable cloud conditions (i.e., irregular lighting due to transient clouds) and high SZAs during late afternoon albedo transect collections was made. Key et al. (2001) reported a 4-6% increase in albedo, on average, under cloudy conditions. Albedo readings have also been reported as unreliable at SZAs beyond 70°, due to an increase in diffuse radiation reaching the ice surface (Schaaf et al., 2011; Stroeve et al., 2005; Stroeve et al., 2013; Wang et al., 2012).

As a proxy for cloud cover, relative cloud cover, hereafter CC, was calculated every second as the ratio of modeled clear-sky and observed incoming solar radiation similar to Box (1997). 'Clear-sky' incoming shortwave fluxes at the surface were calculated with a solar radiance model (Iqbal, 1988). Model inputs of water vapor content, surface pressure, aerosol optical depth at 380 and 500 nm, and area optical thickness were estimated from the Kangerlussuaq AEROsol Robotic NETwork (AERONET) station (Holben et al., 2001). SZA was also modeled with the solar radiance model using latitude, longitude, time of day, and day of year at the Base Met Station. α_{ASD} collected under high CC variability and SZAs approaching extreme angles were subsequently removed. Filtering α_{ASD} data under these criteria ensured the production of a high-quality dataset necessary for subsequent analysis.

Cloud cover and radiative conditions varied among transects (Fig. A1). The majority of α_{ASD} measurements were made at small SZAs (~10:30-12:00 local time), except on 21 and 24 June, when observations were made in late afternoon (15:30-16:30 and 16:40-17:50 local time, respectively). Incoming solar radiation fluxes exhibited considerable range of diurnal variability (average 662 ± 83 W m^{-2}). Outgoing solar radiation displayed similar range of variability at lower magnitudes (average 239 ± 18 W m^{-2}) during transect dates. Derived CC reveals daily range in cloud fractions roughly consistent with incoming solar radiation observations, which on average, remained low (~0.13) indicating that the majority of the transect measurements were collected during nearly cloud-free conditions. During transect times, half-hourly α_{base} changed linearly with SZA, yet remained fairly stable (Fig. A2a). Above 80° SZA, half hourly α_{base} variability increased, confirming that 70° SZA was a suitable threshold for daily average albedo calculations. Installation tilt, heterogeneous and changing surface conditions likely contributed locally to "unstable" α_{base} observations at higher SZAs. A hysteresis observed in α_{top} observations (data not shown) is attributed

primarily to a low installation height (0.5 m), but may also be partly due to changing surface conditions. These effects can compromise the accurate representation of illumination and viewing geometries, resulting in reduced albedo estimates at high SZAs (Kuhn, 1974; Wang et al., 2012; Dumont et al., 2012). As such, Top Met Station measurements, and α_{base} at SZAs greater than 70°, were excluded for most analyses. Despite its limitations, α_{top} were used for α_{ASD} comparison described below.

High range of CC variability, instead of consistently high CC, was found to be responsible for saturating α_{ASD} readings on 17, 21, and 24 June (Fig. A2b). Continuous recalibration of the ASD instrument on 17 and 24 June was inadequate to overcome variable lighting conditions resulting in saturated α_{ASD} readings (> 1). During 21 June, α_{ASD} data did not saturate despite variable sky conditions (0.01-0.52 CC range). Variable cloud conditions on 17, 21, and 24 of June effectively increased the amount of downwelling longwave radiation relative to shortwave radiation available at the surface, of which, the net effect results in a larger portion of solar radiation available to be reflected by the ice surface (Grenfell and Perovich, 2004; Román et al., 2010; Wang et al., 2012). This can translate to an increase in spectral albedo estimates by ~0.06 over active melting ice surfaces (Grenfell and Perovich, 2004).

By removing the majority of shortcomings and uncertainties identified in transect radiative and surface conditions, a high-quality albedo dataset was produced. Optimal SZA, CC, and radiative conditions were observed for 16, 19 and 25 June. α_{ASD} data collected on 17, 21, and 24 June were identified as low-quality based on their dependence on SZA, CC variability, and issues with albedo saturation, and subsequently removed from further analysis (Fig. A2). The first and last high-quality α_{ASD} measurements closest to the AWSs were compared and reveal that they agree reasonably well with α_{base} and α_{top} data (Fig. A3). As much as 62% of α_{ASD} variance is explained by α_{base} and α_{top} ($\alpha_{ASD} = 0.27\alpha_{MET} + 0.46$, where α_{MET} is α_{base} and α_{top} combined). The discrepancy is likely due to differences in exact sample locations and instrumentation. Table 1 provides summary statistics related to high-quality α_{ASD} and transect conditions.

7.3 Installation of meteorological stations

The Top Met Station was installed upon a homogenous clean ice surface, and the Base Met Station was installed above a heterogeneous surface of mixed clean and dirty ice. Both stations measured solar radiation fluxes every 0.5 h at 300-1100 nm, using S-LIB-M003 silicon pyranometers and a U30 data logger (Table 1A; $\pm$ 5% or 10 W m^{-2} precision; Onset Computer Corp., 2010) from 8-26 June. Sensors were attached to a pole drilled into the ice at 1.5 m above the surface, and were kept relatively constant at this height, but occasionally tilted off-level. After a period of heavy melting, the Top Met Station was re-drilled and installed at 0.5 m height and remained at this height as melting seized. A very large hysteresis in α_{top} as a function of SZA suggests that the low installation height resulted in α_{top} errors due to a disproportionally large influence of surface roughness on its measurements. Despite not having observed tilt information for the AWSs, we use a theoretical tilt (for Fig. A2b) in Van den Broeke et al. (2004) to provide a reasonable uncertainty range. Assuming a tilt of 1º on 18 January at Kohnen station, Antarctica (75ºS, 0º) is associated with ~15 W m^{-2} offset in net shortwave at noon local time. This is associated with an absolute error of 5% with a tilt of 1º. Here, we assume double the uncertainty ($\pm$ 10%).
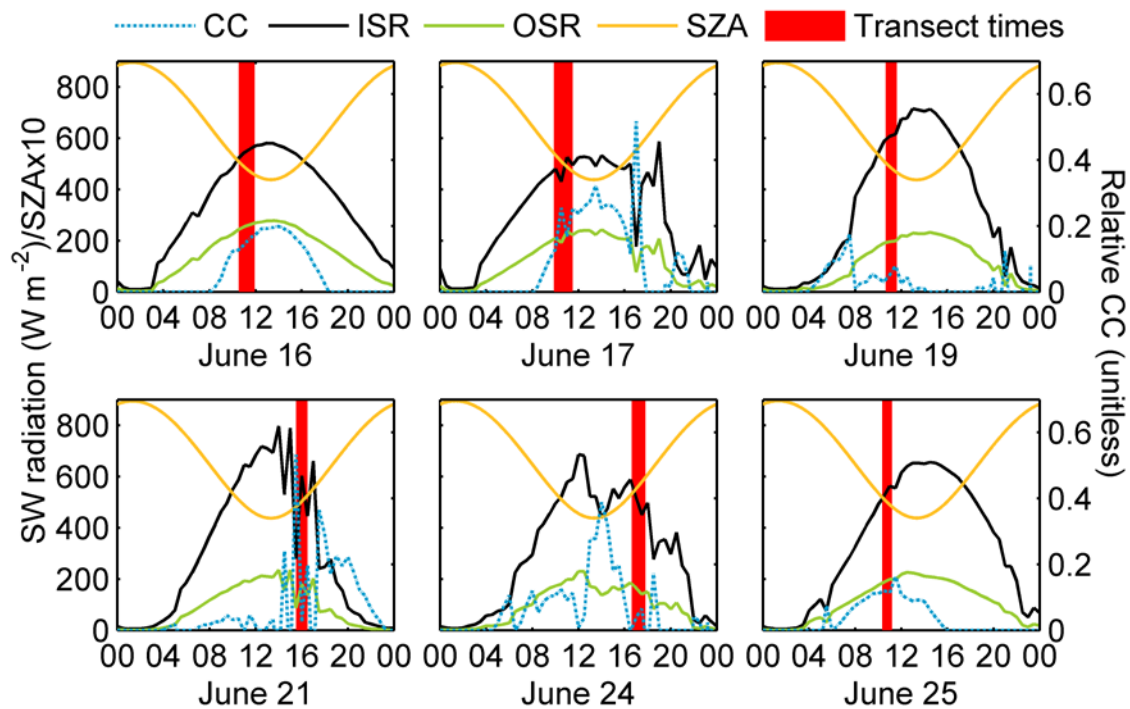
Figure A1. Radiative conditions during transect dates at the Base Met Station, including incoming solar radiation (ISR, black line), outgoing solar radiation (OSR, green line; left y-axis), modeled relative cloud cover (CC, blue stippled line; right y-axis), and solar zenith angles (SZA, yellow line right axis). Red shaded regions show α_{ASD} data collection times.
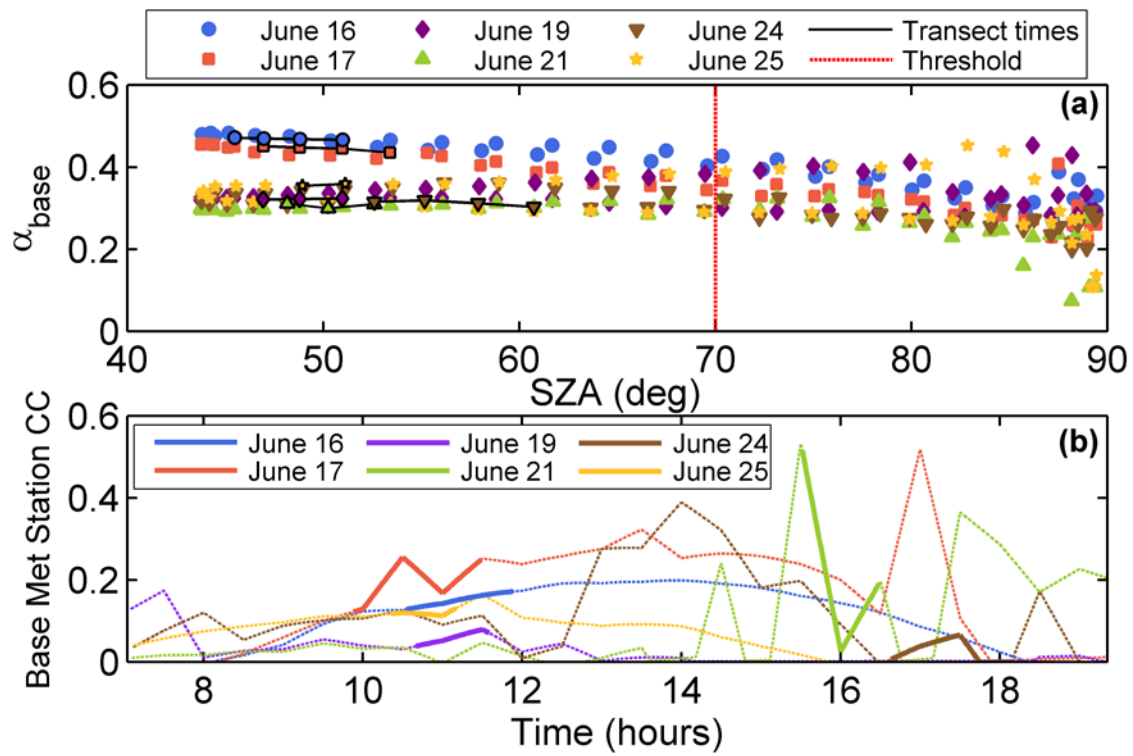
Figure A2. (a) Half hourly broadband α_{base} measurements as a function of SZA. Symbols and colors correspond to transect dates. Transect times correspond to the black line. A SZA threshold at 70° is represented by the red stipple line. (b) Relative CC determined at α_{base} as a function of time during transect dates. Symbols and colors correspond to transect dates. Transect times correspond to bold lines.
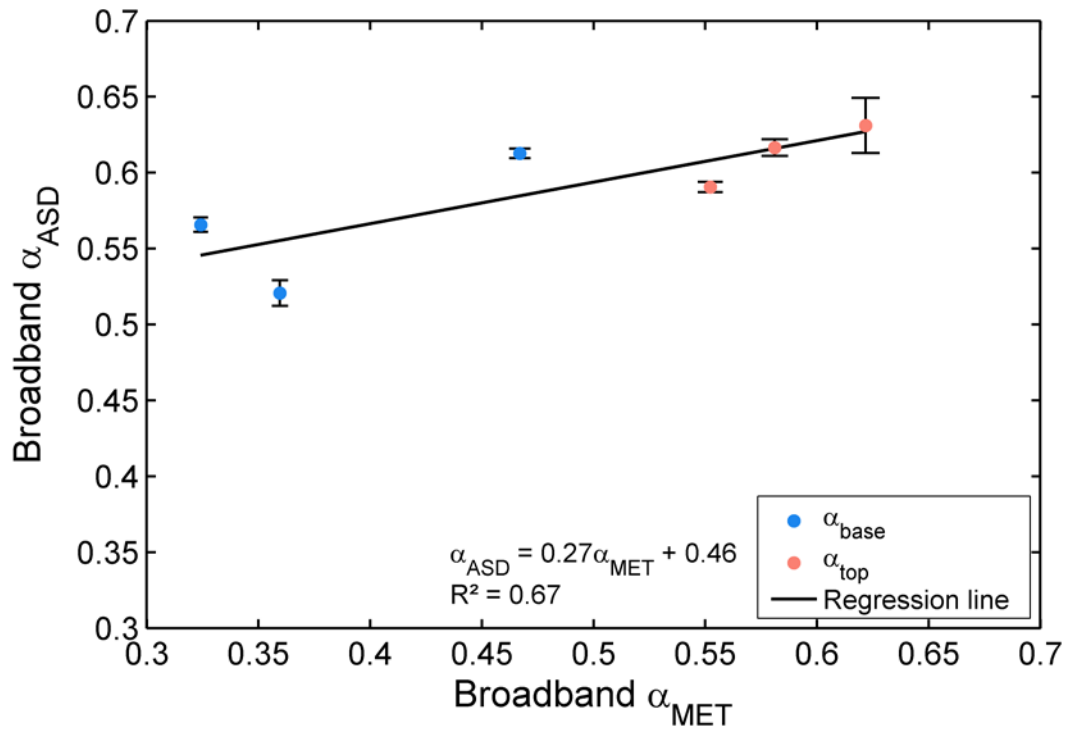
Figure A3. Broadband α_{base} (blue dots) and α_{top} (pink dots) vs. α_{ASD} and α_{MET} (i.e., both α_{base} and α_{top}) measurements fitted to a linear regression equation ($R^2 = 0.67$). The value of α_{ASD} error is based on the standard deviation of individual α_{ASD} measurements.

1   Table 1A. Meteorological station sites and associated variables.

Site	Latitude	Longitude	Elevation (m)	Start Date	End Date
Base Met Station	67.151629	50.027993	511.3	8-Jun	26-Jun
Top Met Station	67.146857	50.001186	586.0	14-Jun	26-Jun

2